

# Group\_inflation\_comparison

February 11, 2025

```
[6]: from inflation_analysis import grouping, calculate_price_indexes
from tabulate import tabulate
from tqdm import tqdm
import matplotlib.pyplot as plt
import pandas as pd

# Parameters
start_year = 2019
end_year = 2022
data_folder="/Users/roykisluk/Downloads/Consumer_Expenditure_Survey/"
top_n = 10
base_year = start_year
years=range(start_year, end_year+1)

groups, total_mmb = grouping(start_year, end_year, cex_data_folder = "
↳data_folder)
groups_mmb = {key: {} for key in groups.keys()}
for key in groups:
    for year in years:
        groups_mmb[key][year] = groups[key][year][['misparmb']]

group_analysis = {}
for key in groups.keys():
    group_number = list(groups.keys()).index(key) + 1
    total_groups = len(groups)
    print(f"Group {group_number}/{total_groups} ({key}) started.")
    combined_df, combined_secondary_df, combined_primary_df, yearly_price_index,
↳ calculate_price_indexes(
        start_year, end_year, base_year, group_mmb=groups_mmb[key],
↳cex_data_folder=data_folder, verbose=False
    )
    group_analysis[key] = {
        'combined_secondary_df': combined_secondary_df,
        'combined_primary_df': combined_primary_df,
        'yearly_price_index': yearly_price_index
    }
```

```
print(f"Group {group_number}/{total_groups} ({key}) successfully computed.")
```

Group 1/11 (Arabs) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.46it/s]

Calculating price indexes: 100%| | 4/4 [00:03<00:00, 1.12it/s]

Group 1/11 (Arabs) successfully computed.

Group 2/11 (Haredi) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.85it/s]

Calculating price indexes: 100%| | 4/4 [00:03<00:00, 1.25it/s]

Group 2/11 (Haredi) successfully computed.

Group 3/11 (Low\_inc) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.81it/s]

Calculating price indexes: 100%| | 4/4 [00:02<00:00, 1.34it/s]

Group 3/11 (Low\_inc) successfully computed.

Group 4/11 (High\_inc) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.85it/s]

Calculating price indexes: 100%| | 4/4 [00:05<00:00, 1.30s/it]

Group 4/11 (High\_inc) successfully computed.

Group 5/11 (Young) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.67it/s]

Calculating price indexes: 100%| | 4/4 [00:03<00:00, 1.02it/s]

Group 5/11 (Young) successfully computed.

Group 6/11 (Old) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.71it/s]

Calculating price indexes: 100%| | 4/4 [00:04<00:00, 1.23s/it]

Group 6/11 (Old) successfully computed.

Group 7/11 (Low\_SES) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.80it/s]

Calculating price indexes: 100%| | 4/4 [00:04<00:00, 1.08s/it]

Group 7/11 (Low\_SES) successfully computed.

Group 8/11 (High\_SES) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.84it/s]

Calculating price indexes: 100%| | 4/4 [00:02<00:00, 1.74it/s]

Group 8/11 (High\_SES) successfully computed.

Group 9/11 (Muslim) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.78it/s]

Calculating price indexes: 100%| | 4/4 [00:03<00:00, 1.25it/s]

Group 9/11 (Muslim) successfully computed.

Group 10/11 (Christian) started.

Loading price data: 100%| | 4/4 [00:02<00:00, 1.83it/s]  
 Calculating price indexes: 100%| | 4/4 [00:02<00:00, 1.86it/s]  
 Group 10/11 (Christian) successfully computed.  
 Group 11/11 (Druze) started.  
 Loading price data: 100%| | 4/4 [00:02<00:00, 1.84it/s]  
 Calculating price indexes: 100%| | 4/4 [00:01<00:00, 2.66it/s]  
 Group 11/11 (Druze) successfully computed.

```
[2]: gen_pop_df, gen_pop_secondary_df, gen_pop_primary_df,
      ↪gen_pop_yearly_price_index = calculate_price_indexes(start_year, end_year,
      ↪base_year, cex_data_folder=data_folder, verbose=False)
```

Loading price data: 100%| | 8/8 [00:08<00:00, 1.00s/it]  
 Calculating price indexes: 100%| | 8/8 [00:33<00:00, 4.14s/it]

```
[3]: group_counts = {group: {year: len(groups_mmb[group][year]) for year in
      ↪groups_mmb[group]} for group in groups_mmb}
      # Create a dataframe with number of observations per year per group
      observations_df = pd.DataFrame(group_counts).T

      # Calculate the relative share of each group per year
      total_observations_per_year = observations_df.sum(axis=0)
      relative_share_df = observations_df.div(total_observations_per_year, axis=1) *
      ↪100

      # Combine the absolute and relative values into a single dataframe
      combined_df = observations_df.join(relative_share_df, rsuffix='_share')

      # Display the dataframe
      print(tabulate(combined_df, headers='keys', tablefmt='psql'))
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
|          | 2015 | 2016 | 2017 | 2018 | 2019 | 2020 | 2021 |
2022 | 2015_share | 2016_share | 2017_share | 2018_share | 2019_share
| 2020_share | 2021_share | 2022_share |
+-----+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
| Arabs    | 1136 | 1273 | 1327 | 1145 | 1103 | 513 | 951 |
727 | 11.5236 | 11.9351 | 12.2825 | 11.1295 | 10.9273 |
7.49014 | 11.1476 | 9.51944 |
| Haredi   | 734 | 778 | 757 | 786 | 565 | 440 | 551 |
595 | 7.44573 | 7.29421 | 7.00666 | 7.63997 | 5.59738 |
```

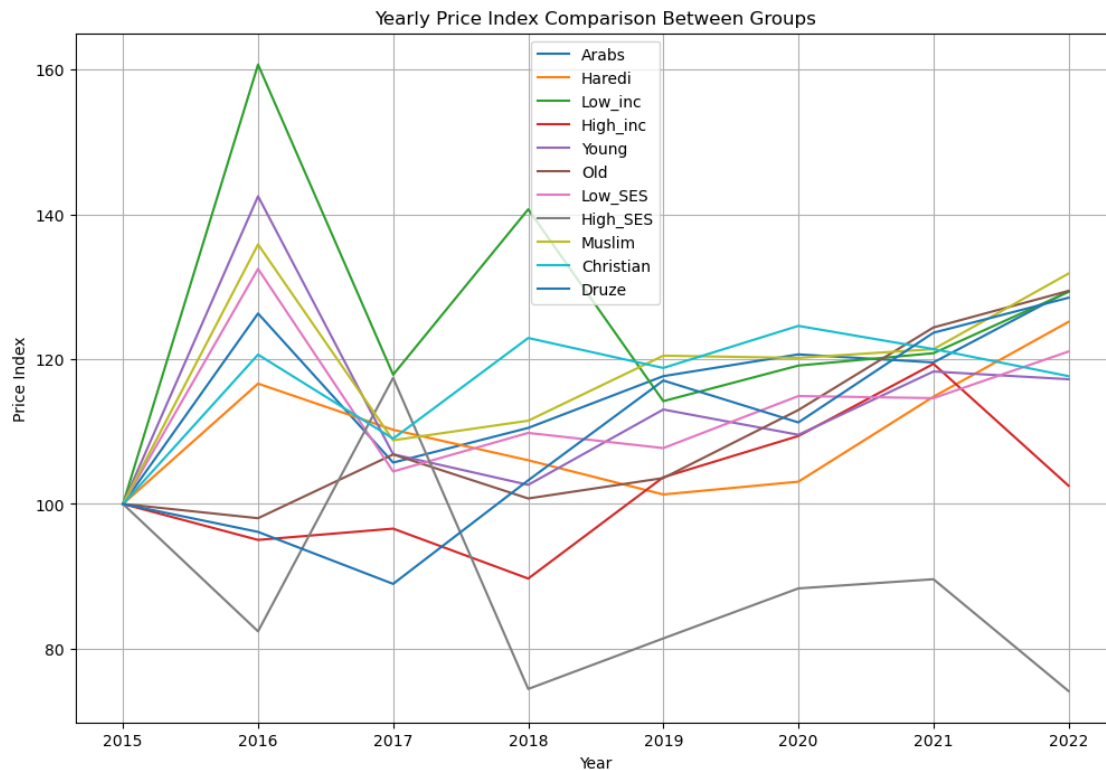
6.4243	6.4588	7.79102						
Low_inc	873	890	940	880	646	397	503	
515	8.85575	8.34427	8.70048	8.55365	6.39984			
5.79647	5.89614	6.74349						
High_inc	1667	1770	1741	1766	1848	1432	1479	
1162	16.9101	16.5948	16.1144	17.1656	18.3079			
20.9082	17.3368	15.2154						
Young	1340	1354	1358	1278	1108	718	877	
820	13.593	12.6945	12.5694	12.4222	10.9768			
10.4833	10.2802	10.7372						
Old	2091	2372	2348	2375	2279	1786	1779	
1663	21.2112	22.2389	21.7327	23.0851	22.5778			
26.0768	20.8534	21.7756						
Low_SES	616	695	728	654	1146	807	1204	
1211	6.24873	6.51603	6.73825	6.35692	11.3533			
11.7827	14.1132	15.857						
High_SES	121	95	102	127	189	163	153	
146	1.22743	0.890681	0.944095	1.23445	1.8724			
2.37991	1.79346	1.91175						
Muslim	880	999	1056	920	870	381	710	
583	8.92676	9.36621	9.77416	8.94246	8.61898			
5.56286	8.32259	7.63389						
Christian	260	309	300	234	231	151	205	
137	2.63745	2.89706	2.77675	2.27449	2.28849			
2.2047	2.403	1.7939						
Druze	140	131	147	123	109	61	119	
78	1.42017	1.2282	1.36061	1.19557	1.07985			
0.890641	1.39491	1.02134						

```
[4]: # Extract yearly price indexes for each group
group_yearly_price_indexes = {group:
    ↪group_analysis[group]['yearly_price_index'] for group in group_analysis}

# Plot the yearly price indexes
plt.figure(figsize=(12, 8))
for group, price_indexes in group_yearly_price_indexes.items():
    years = list(price_indexes.keys())
    indexes = list(price_indexes.values())
    plt.plot(years, indexes, label=group)

plt.xlabel('Year')
plt.ylabel('Price Index')
plt.title('Yearly Price Index Comparison Between Groups')
plt.legend()
```

```
plt.grid(True)
plt.show()
```



```
[7]: # Extract yearly price indexes for each group including general population
group_yearly_price_indexes = {group:
    ↪group_analysis[group]['yearly_price_index'] for group in group_analysis}
group_yearly_price_indexes['All'] = gen_pop_yearly_price_index

# Define colors for each group
colors = ['#1f77b4', '#ff7f0e', '#2ca02c', '#d62728', '#9467bd', '#8c564b',
    ↪'#e377c2', '#7f7f7f', '#bcbd22', '#17becf', '#9edae5', '#c7c7c7']

# Create subplots for each year
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(15, 10), sharey=True)
axes = axes.flatten()

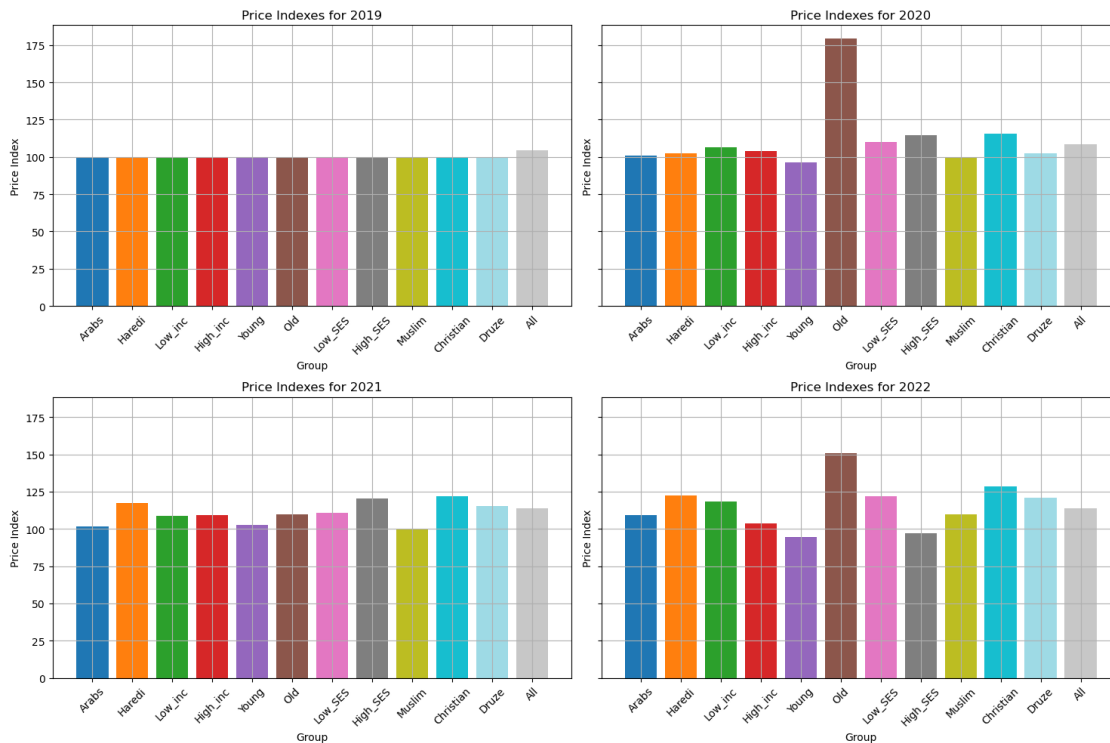
for i, year in enumerate(years[:len(axes)]):
    ax = axes[i]
    groups = list(group_yearly_price_indexes.keys())
    price_indexes = [group_yearly_price_indexes[group][year] for group in
    ↪groups]
    ax.bar(groups, price_indexes, color=colors)
```

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ax.set_title(f'Price Indexes for {year}')
ax.set_xlabel('Group')
ax.set_ylabel('Price Index')
ax.grid(True)
ax.tick_params(axis='x', rotation=45)

plt.tight_layout()
plt.show()

```



```

[8]: fig, axes = plt.subplots(nrows=4, ncols=3, figsize=(20, 15), sharey=True)
axes = axes.flatten()

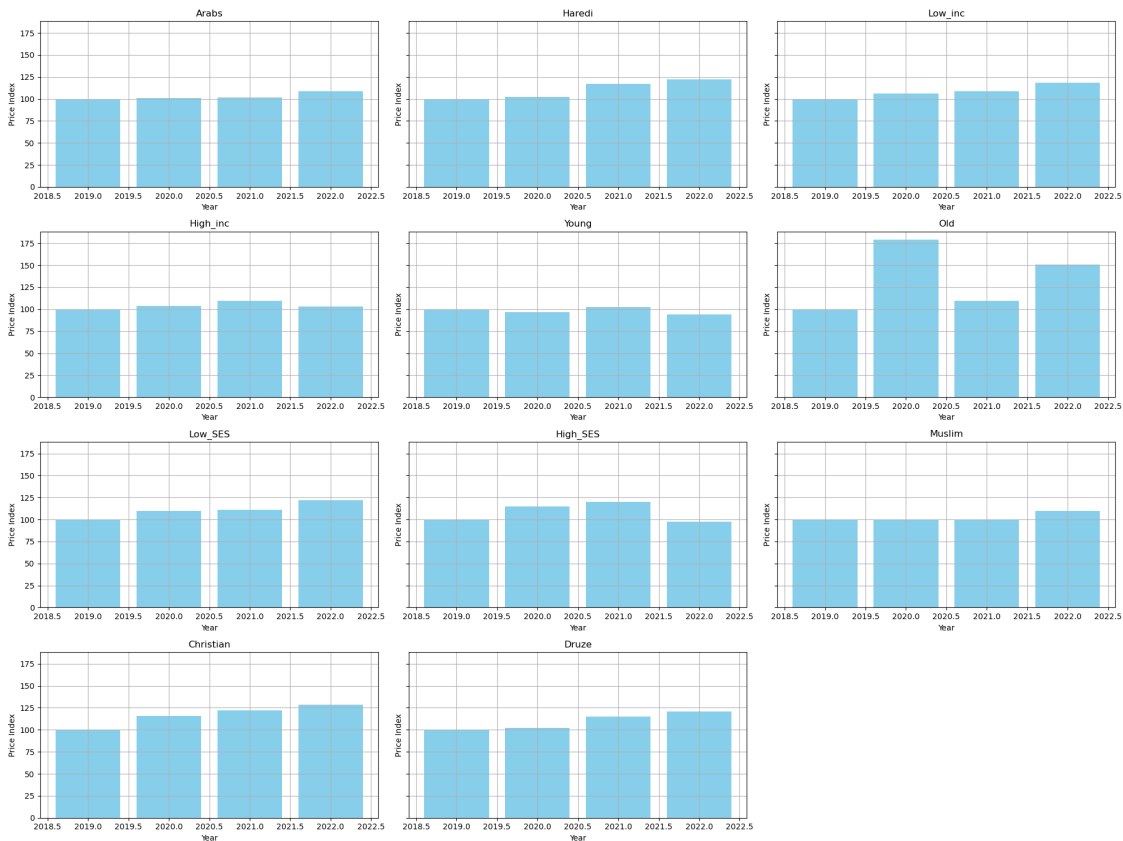
for i, (group, data) in enumerate(group_analysis.items()):
    yearly_price_index = data['yearly_price_index']
    years = list(yearly_price_index.keys())
    indexes = list(yearly_price_index.values())

    axes[i].bar(years, indexes, color='skyblue')
    axes[i].set_title(group)
    axes[i].set_xlabel('Year')
    axes[i].set_ylabel('Price Index')
    axes[i].grid(True)

```

```
# Remove any empty subplots
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()
```



```
[9]: for group, analysis in group_analysis.items():
    # Calculate the weight differences between the group and the general
    ↪ population for secondary categories
    weight_diff_df = analysis['combined_secondary_df'].copy()
    weight_diff_df['weight_diff'] = weight_diff_df['weight'] -
    ↪ gen_pop_secondary_df['weight']

    # Sort by the absolute value of the weight differences in descending order
    weight_diff_df['abs_weight_diff'] = weight_diff_df['weight_diff'].abs()
    sorted_weight_diff_df = weight_diff_df.sort_values(by='abs_weight_diff',
    ↪ ascending=False)

    # Display the top_n largest gaps using tabulate
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print(f"Group: {group}")
print(tabulate(sorted_weight_diff_df.head(top_n)[['Year', 'prodcode', 'weight', 'description', 'weight_diff']], headers='keys', tablefmt='psql'))
print("\n")

```

Group: Arabs

	Year	prodcode	description	weight	weight_diff
249	2022	395	Personal Jewelry And Watches	0.00278528	-0.131459
54	2019	384	Other Transportation Expenses	0.0063836	-0.126337
116	2020	383	Vehicle Expenses	0.11973	
184	2021	393	Personal Products And Cosmetics	0.0145025	-0.119572
179	2021	383	Vehicle Expenses	0.115379	
53	2019	383	Vehicle Expenses	0.112121	
242	2022	383	Vehicle Expenses	0.106663	
119	2020	391	Cigarettes Tobacco And Smoking Supplies	0.0300511	-0.103199
66	2020	302	Meat And Poultry	0.0924262	0.0855918
129	2021	302	Meat And Poultry	0.0886994	0.08524

Group: Haredi

	Year	prodcode	description	weight	weight_diff
184	2021	395	Personal Jewelry And Watches	0.00416806	-0.129906
54	2019	384	Other Transportation Expenses	0.00360779	-0.129113
119	2020	392	Personal Services And Cosmetics	0.00463257	





Group: High\_inc

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	Year	prodcode	description		
weight	weight_diff				
-----+-----+-----+-----+-----					
---+-----+-----+-----+-----					
246	2022	383	Vehicle Expenses		0.158819
0.153794					
249	2022	391	Cigarettes Tobacco And Smoking Supplies		
0.00483538	-0.129409				
184	2021	385	Mail Telephone And Communication		
0.00645372	-0.127621				
119	2020	384	Other Transportation Expenses		
0.00634826	-0.126902				
137	2021	308	Meals Outside Home		
0.0956662	0.0893899				
182	2021	383	Vehicle Expenses		
0.0974391	0.0838352				
118	2020	383	Vehicle Expenses		
0.0971747	0.0834259				
201	2022	308	Meals Outside Home		
0.0877109	0.0769389				
203	2022	311	Potatoes And Sweet Potatoes		
0.00283584	-0.0441704				
131	2021	302	Meat And Poultry		
0.0481333	0.0412608				
+-----+-----+-----+-----+-----					
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Group: Young

	Year	prodcode	description		
weight	weight_diff				
54	2019	384	Other Transportation Expenses		
0.00469209	-0.128028				
119	2020	385	Mail Telephone And Communication		
0.0111502	-0.1221				
249	2022	392	Personal Services And Cosmetics		
0.0131016	-0.121143				
184	2021	391	Cigarettes Tobacco And Smoking Supplies		
0.0212876	-0.112787				

181	2021	383	Vehicle Expenses	
0.0935405		0.088328		
53	2019	383	Vehicle Expenses	
0.0979747		0.0843083		
117	2020	383	Vehicle Expenses	
0.0974486		0.0839201		
245	2022	383	Vehicle Expenses	
0.0902798		0.0701298		
72	2020	308	Meals Outside Home	
0.0739085		0.067667		
136	2021	308	Meals Outside Home	
0.0701319		0.0593855		
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Group: Old

weight_diff	Year	prodcode	description	weight
-0.131371	2021	384	Other Transportation Expenses	0.00270334
-0.129407	2020	384	Other Transportation Expenses	0.00384282
-0.124995	2022	385	Mail Telephone And Communication	0.0092496
-0.124975	2019	382	Travel Abroad	0.00774532
0.0920758	2019	383	Vehicle Expenses	0.0974318
0.0853741	2021	383	Vehicle Expenses	0.0991997
0.0847057	2022	383	Vehicle Expenses	0.0983421
0.084261	2020	383	Vehicle Expenses	0.0980098
0.0616188	2021	302	Meat And Poultry	0.0684913
0.061075	2022	302	Meat And Poultry	0.067964

Group: Low\_SES

	Year	prodcode	description		weight		
weight_diff							
249	2022	394	Legal And Other Services		0.0047145		
-0.12953							
184	2021	392	Personal Services And Cosmetics		0.00627667		
-0.127798							
119	2020	385	Mail Telephone And Communication		0.0147653		
-0.118484							
117	2020	383	Vehicle Expenses		0.104771		
0.0912424							
243	2022	383	Vehicle Expenses		0.105859		
0.0908215							
180	2021	383	Vehicle Expenses		0.10359		
0.0834932							
193	2022	302	Meat And Poultry		0.0859717		
0.0832089							
128	2021	300	Bread Grains and Pastries		0.0531601		
0.0504126							
191	2022	300	Bread Grains and Pastries		0.0534574		
0.0483045							
197	2022	306	Soft Drinks		0.0127775		
-0.0423986							

Group: High\_SES

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	Year		prodcode	description		weight				
weight_diff		-----+								
	-----+					-----+				
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236	2022		383	Vehicle Expenses		0.230227				
0.227364										
53	2019		383	Vehicle Expenses		0.208694				
0.195027										
119	2020		396	Bags Suitcases And Other Products		0.00305701				
-0.130193										
54	2019		384	Other Transportation Expenses		0.00413969				
-0.128581										
184	2021		398	Expenses Not Elsewhere Specified		0.0120238				
-0.12205										
72	2020		308	Meals Outside Home		0.123332				



weight	weight_diff						
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225	2022	383	Vehicle Expenses		0.159651		
0.154686							
166	2021	383	Vehicle Expenses		0.150521		
0.14903							
51	2019	383	Vehicle Expenses		0.146481		
0.141323							
119	2021	301	Vegetable Oils And Products				
0.00548327	-0.127767						
184	2022	307	Alcoholic Beverages				
0.0122635	-0.121811						
54	2019	391	Cigarettes Tobacco And Smoking Supplies				
0.0221254	-0.110595						
109	2020	383	Vehicle Expenses		0.132531		
0.103045							
64	2020	302	Meat And Poultry				
0.0996246	0.096186						
120	2021	302	Meat And Poultry		0.091822		
0.0864336							
179	2022	302	Meat And Poultry				
0.0916972	0.0844143						
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Group: Druze

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	Year	prodcode	description		weight		
weight_diff							
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207	2022	383	Vehicle Expenses		0.162452		
0.143969							
54	2019	394	Legal And Other Services		0.000122185		
-0.132598							
98	2020	383	Vehicle Expenses		0.151386		
0.129191							
152	2021	383	Vehicle Expenses		0.152783		
0.125564							
48	2019	383	Vehicle Expenses		0.13748		
0.122021							
164	2022	302	Meat And Poultry		0.118103		
0.107798							
60	2020	302	Meat And Poultry		0.112965		
0.107199							

119	2021	312	Fresh Vegetables	0.0282493
-0.105001				
184	2022	336	Municipal Taxes	0.0461439
-0.0879303				
77	2020	332	Electricity Gas And Fuel For Home	0.104869
0.0865322				
+-----+	+-----+	+-----+	+-----+	+-----+
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```
[10]: fig, axes = plt.subplots(nrows=4, ncols=3, figsize=(20, 15), sharey=False)
axes = axes.flatten()

for i, (group, analysis) in enumerate(group_analysis.items()):
    # Calculate the weight differences between the group and the general
    ↪ population for secondary categories
    weight_diff_df = analysis['combined_secondary_df'].copy()
    weight_diff_df['weight_diff'] = weight_diff_df['weight'] -
    ↪ gen_pop_secondary_df['weight']

    # Sort by the absolute value of the weight differences in descending order
    weight_diff_df['abs_weight_diff'] = weight_diff_df['weight_diff'].abs()
    sorted_weight_diff_df = weight_diff_df.sort_values(by='abs_weight_diff',
    ↪ ascending=False)

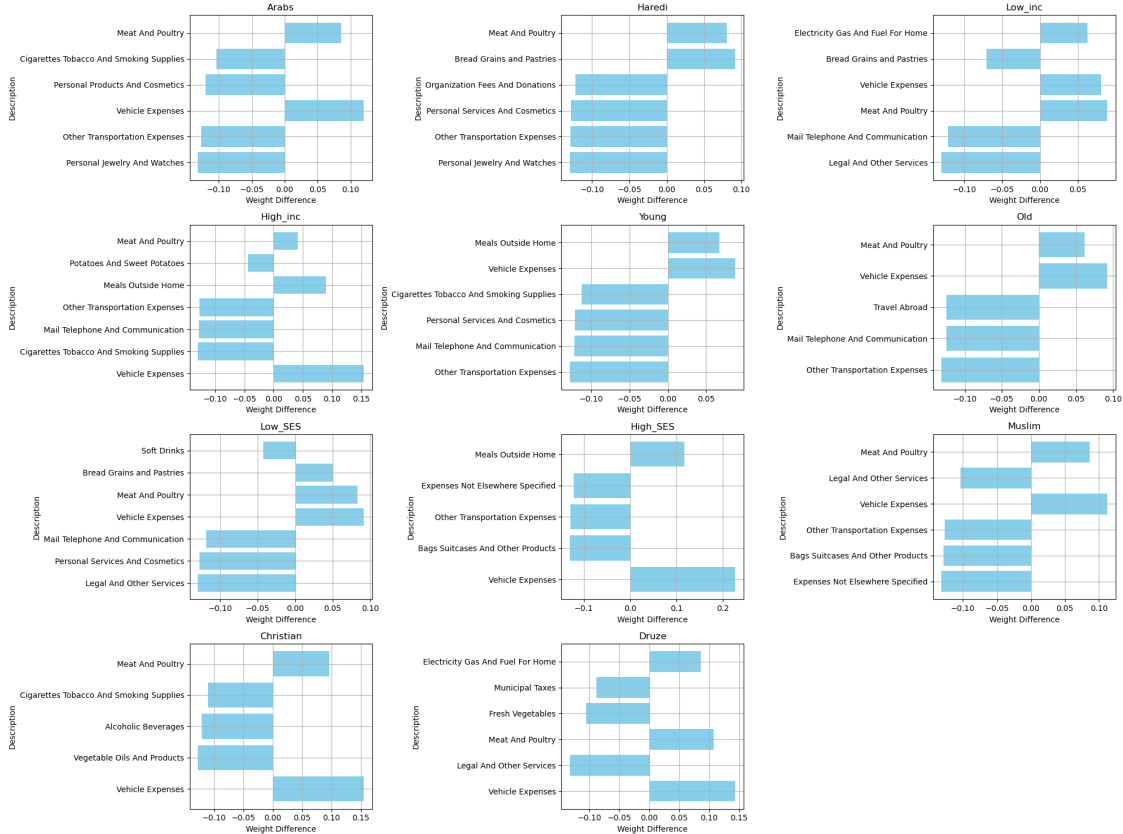
    # Replace NaN values in 'description' with an empty string
    sorted_weight_diff_df['description'] = sorted_weight_diff_df['description'].
    ↪ fillna('')

    # Select the top n largest gaps
    top_n_weight_diff_df = sorted_weight_diff_df.head(top_n)

    # Plot the top n largest gaps
    axes[i].barh(top_n_weight_diff_df['description'],
    ↪ top_n_weight_diff_df['weight_diff'], color='skyblue')
    axes[i].set_title(group)
    axes[i].set_xlabel('Weight Difference')
    axes[i].set_ylabel('Description')
    axes[i].grid(True)

# Remove any empty subplots
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()
```



```
[11]: fig, axes = plt.subplots(nrows=4, ncols=3, figsize=(20, 15), sharey=False)
axes = axes.flatten()

for i, (group, analysis) in enumerate(group_analysis.items()):
    secondary_df = analysis['combined_secondary_df']
    # Filter for increases in price indexes
    increased_price_df = secondary_df[secondary_df['price_index'] > 100]
    # Calculate the contribution to the yearly price index
    increased_price_df = increased_price_df.copy() # Ensure you're working
    ↪with a copy
    increased_price_df.loc[:, 'contribution'] =
    ↪increased_price_df['price_index'] * increased_price_df['weight']
    # Sort by contribution in descending order
    top_contributors = increased_price_df.sort_values(by='contribution',
    ↪ascending=False).head(top_n)

    # Replace NaN values in 'description' with a placeholder text
    top_contributors['description'] = top_contributors['description'].
    ↪fillna('No Description')
```



```

# Plot the top contributors
axes[i].barh(top_contributors['description'],
top_contributors['contribution'], color='skyblue')
axes[i].set_title(group)
axes[i].set_xlabel('Contribution to Price Index')
axes[i].set_ylabel('Description')
axes[i].grid(True)

# Remove any empty subplots
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()

```

